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Supplemental Material

Acute Impact of Hourly Ambient Air Pollution on Preterm Birth

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CO: carbon monoxide; NO₂: nitrogen dioxide; OR: odds ratio; PM_{2.5}: particulate matter 2.5 micrometres or less in diameter; PM₁₀: particulate matter 10 micrometres or less in diameter; ppb: parts per billion; SO₂: sulphur dioxide; µg/m³: microgram per cubic meter.

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R codes

```
#####
##### R codes for case-crossover design matching #####
##### case and control by the same hour of the same #####
##### day of the week in the same month #####
#####

#####
##### Be aware that this is simulated data. It is #####
##### not the same as the study data #####
#####

#####
##### simulate hourly data for one year #####
#####
#DateTime=format(seq(as.POSIXct("2009-01-01"), by=(+1 hour),
#                   length.out=365*24), format="%Y-%m-%d-%H") ## date hour
date<-format(as.POSIXct(DateTime, format="%Y-%m-%d-%H"), "%Y-%m-%d") ## date
hour<-format(as.POSIXct(DateTime, format="%Y-%m-%d-%H"), "%H") ## hour
month<-format(as.POSIXct(DateTime, format="%Y-%m-%d-%H"), "%m") ## month
dow<-format(as.POSIXct(DateTime, format="%Y-%m-%d-%H"), "%w") ## day of the week
ymdowhour<-format(as.POSIXct(DateTime, format="%Y-%m-%d-%H"),
                  "%Y-%m-%w-%H") ## year-month-dow-hour

#####
##### create a data set for air pollution
data<-data.frame(DateTime,date,hour,month,dow,ymdowhour)
data$NO2<-rnorm(nrow(data),mean=10, sd=5) ## create NO2
library(tsModel) ## package for creating moving average of air pollution
data$NO2_lag0_24<-runMean(data$NO2,0:24) ## 0-24 hours' average of NO2

#####
##### create a data set for preterm birth (PB)
##### not all hours have PB
data.PB<-data[sample(nrow(data),1000),] ## sample 1000 hours which have preterm
birth
data.PB$sex<-sample(c("male","female"),1000,replace=T) ## create sex variable
data.PB$smoking<-sample(c("smoke","nosmoke"),1000,replace=T) ## create smoking
variable

#####
##### match case and control
for (i in 1:nrow(data.PB)){
  ymdowh_cc<-data.PB[i,]$ymdowhour
  date_case<-data.PB[i,]$date
  data_cc<-subset(data,data$ymdowhour%in%ymdowh_cc)
  data_cc$case<-as.numeric(data_cc$date%in%date_case)
  data_cc$time<-as.numeric(!data_cc$date%in%date_case)
  data_cc$strata<-i
  data_cc$sex<-data.PB[i,]$sex
  data_cc$smoking<-data.PB[i,]$smoking
  if (i==1){final.data=data_cc}else{final.data=rbind(final.data,data_cc)}
}

#####
##### start to perform conditional logistic regression #####
#####
```

```
library(survival)

### fit linear model
modell<-coxph(Surv(time,case)~NO2_lag0_24+strata(strata),
               method = c("breslow"),data=final.data)
summary(modell)
AIC(modell)

### fit non-linear model with natural cubic spline
library(dlnm)
cb.no2<-onebasis(final.data$NO2_lag0_24,fun="ns",df=4)
model2<-coxph(Surv(time,case)~cb.no2+strata(strata),
               method = c("breslow"),data=final.data)
summary(model2)
AIC(model2)
pred.no2<-crosspred(cb.no2,model2)
plot(pred.no2)
```

Table S1: Summary statistics of daily air pollution and weather conditions during 2009 – 2013 in Brisbane, Australia.

Variables	Mean ± SD	Percentiles				
		5th	25th	50th	75th	95th
PM2.5 ($\mu\text{g}/\text{m}^3$)	6.32 ± 3.88	3.28	4.41	5.59	7.23	10.93
PM10 ($\mu\text{g}/\text{m}^3$)	17.27 ± 14.23	9.96	13.18	15.76	18.92	25.94
NO2 (ppb)	6.52 ± 2.57	2.93	4.55	6.06	8.34	11.25
SO2 (ppb)	1.95 ± 1.38	0.71	1.15	1.53	2.18	4.77
O3 (ppb)	17.27 ± 4.56	10.26	14.06	16.99	20.1	25.12
CO (ppb)	219.25 ± 115.08	62.55	125.83	200.00	300.78	417.03
Temperature (°C)	21.97 ± 3.69	15.67	18.92	22.75	25.11	26.94
Relative humidity (%)	70.82 ± 8.50	55.43	66.26	71.18	76.24	83.94

CO: carbon monoxide; NO2: nitrogen dioxide; O3: ozone; PM2.5: particulate matter 2.5 micrometres or less in diameter; PM10: particulate matter 10 micrometres or less in diameter; ppb: parts per billion; SD: standard deviation; SO2: sulphur dioxide; $\mu\text{g}/\text{m}^3$: microgram per cubic meter.

Table S2: Pearson correlation coefficients for hourly air pollution and weather conditions.

	PM2.5	PM10	NO2	SO2	O3	CO	Temperature	Humidity
PM2.5	1.00							
PM10	0.71**	1.00						
NO2	0.20**	0.02**	1.00					
SO2	0.03**	0.01*	-0.03**	1.00				
O3	0.19**	0.12**	-0.34**	0.18**	1.00			
CO	0.18**	0.02**	0.43**	0.02**	-0.12**	1.00		
Temperature	-0.07**	0.04**	-0.57**	0.18**	0.45**	-0.22**	1.00	
Humidity	-0.16**	-0.17**	0.18**	-0.13**	-0.73**	0.10**	-0.37**	1.00

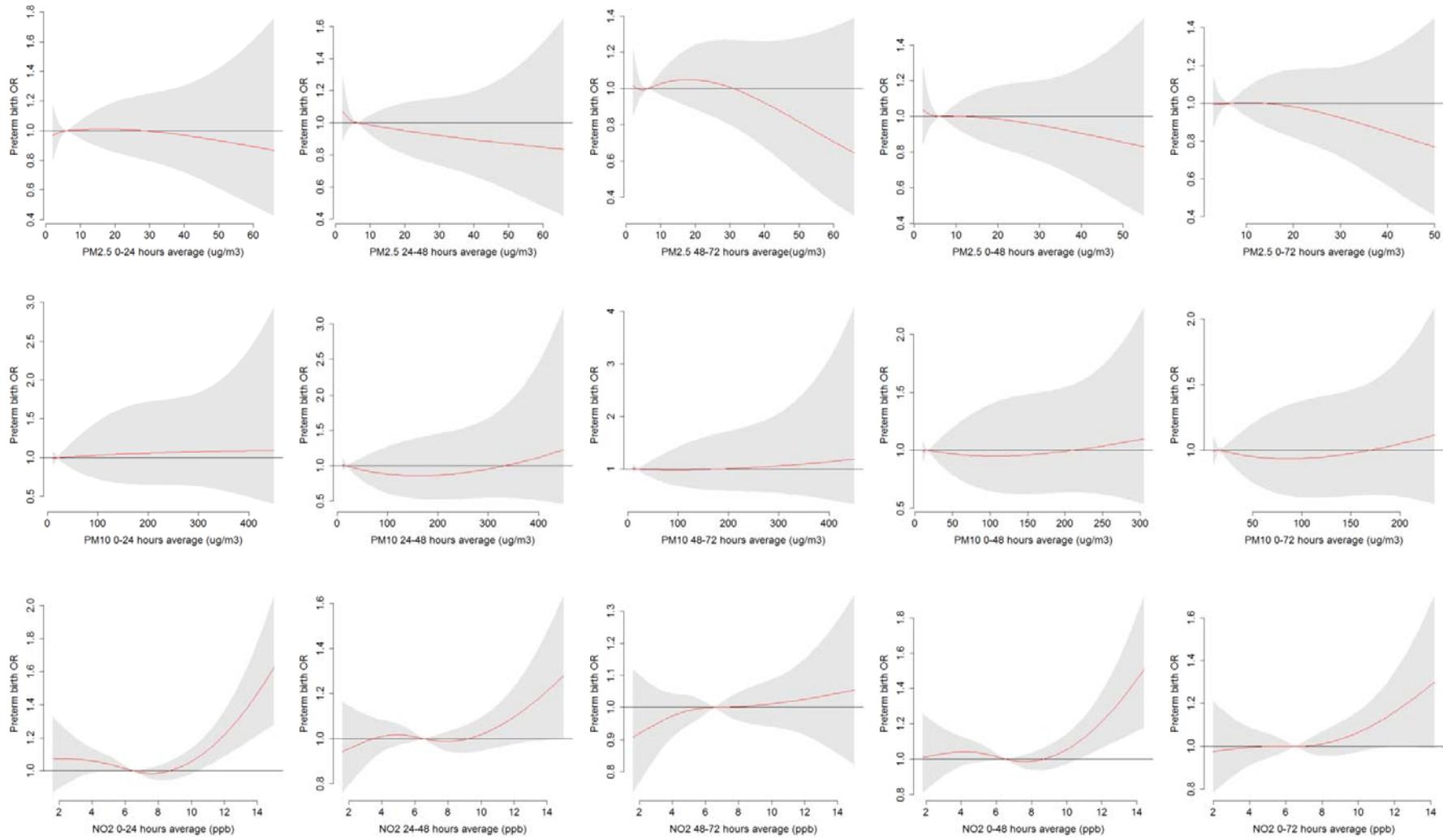
*p < 0.05; **p < 0.01.

CO: carbon monoxide; NO2: nitrogen dioxide; O3: ozone; PM2.5: particulate matter 2.5 micrometres or less in diameter; PM10: particulate matter 10 micrometres or less in diameter; SO2: sulphur dioxide.

Table S3: P-values for differences for the effect estimates by level of demographic factors, corresponding to Table 5.

Factors	p-value for difference			
	NO2	SO2	CO	
Maternal age (years)				
< 35 VS ≥ 35	75th VS Threshold 95th VS Threshold	0.69 0.30	0.61 0.69	0.79 0.31
Pre-pregnancy medical conditions				
No VS Yes	75th VS Threshold 95th VS Threshold	0.03 0.05	0.69 0.52	0.98 0.78
Previous pregnancy				
0 VS ≥ 1	75th VS Threshold 95th VS Threshold	0.53 0.07	0.13 0.79	0.13 0.19
Smoking				
Yes VS No	75th VS Threshold 95th VS Threshold	0.04 0.04	0.52 0.42	0.27 0.24
Number of births				
Single VS Multiple	75th VS Threshold 95th VS Threshold	0.38 0.08	0.15 0.02	0.84 0.33
Baby gender				
Female VS Male	75th VS Threshold 95th VS Threshold	0.65 0.61	0.13 0.76	0.31 0.36
Social-economic level				
Index 1 – 5 VS Index 6 – 10	75th VS Threshold 95th VS Threshold	0.64 0.94	0.76 0.32	<0.0001 0.002

CO: carbon monoxide; NO2: nitrogen dioxide; SO2: sulphur dioxide.



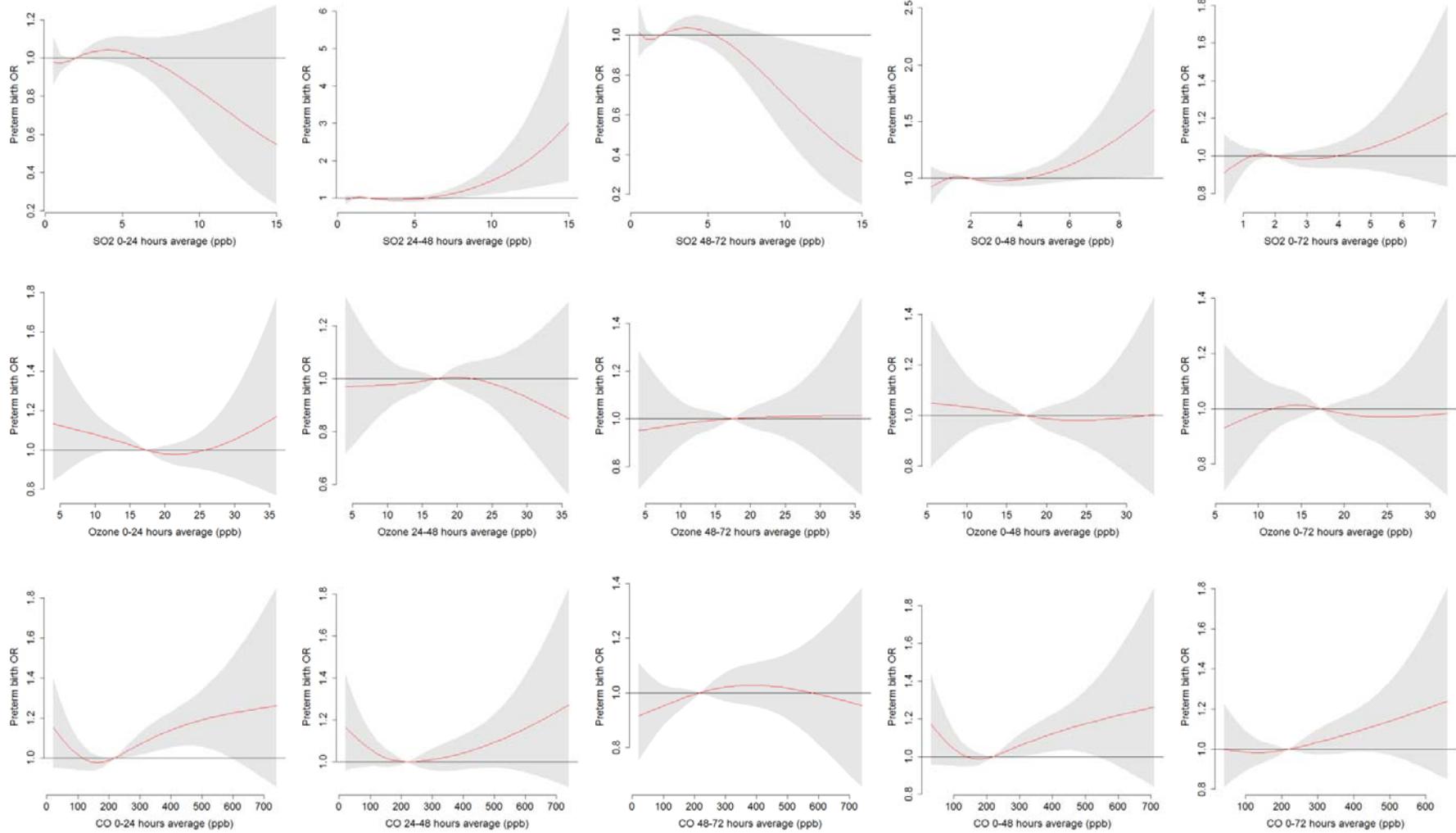


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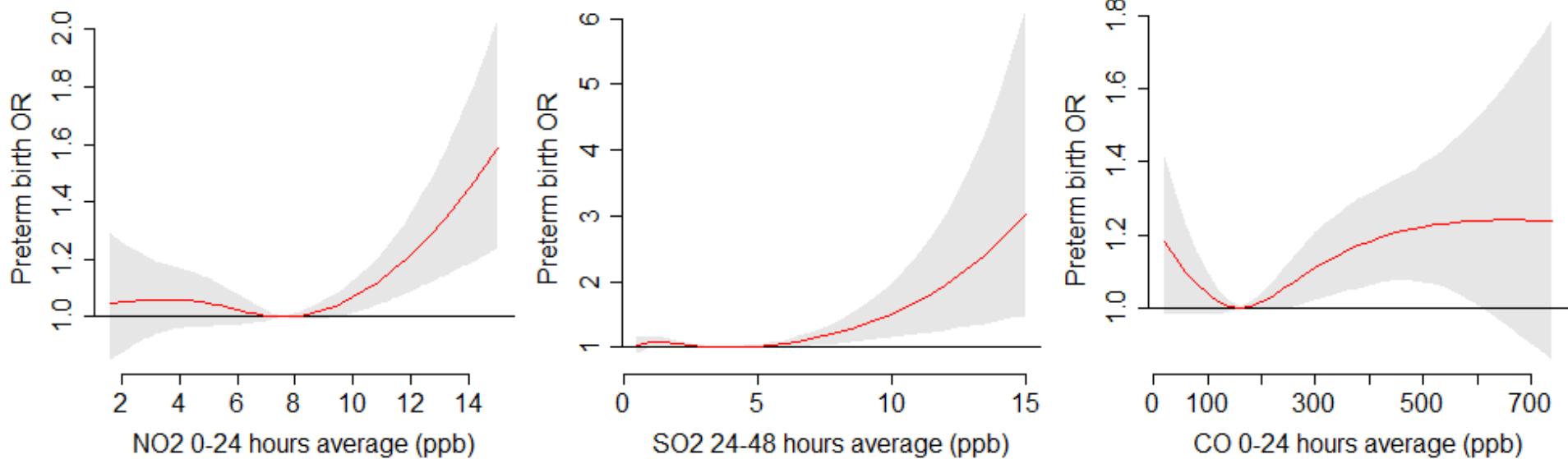


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